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## Knowledge-Based Asynchronous Programming

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**Abstract.** A knowledge-based program is a high-level description of the behaviour of agents in terms of knowledge that an agent must have before (s)he may perform an action. The definition of the semantics of knowledge-based programs is problematic, since it involves a vicious circle; the knowledge of an agent is defined in terms of the possible behaviours of the program, while the possible behaviours are determined by the actions which depend on knowledge. We define the semantics of knowledge-based programs via an iteration approach generalizing the well-known fixpoint construction. We propose a specific iteration as the semantics of a knowledge-based program, and justify our choice by a number of examples, including the Unexpected Hanging Paradox.

**Keywords:** knowledge-based programming, semantics of programming languages, concurrent programming, asynchronicity, unexpected hanging paradox.

### 1. Introduction

A knowledge-based program (KBP) is a program with explicit tests for knowledge. KBPs can be used as high-level descriptions of the behaviour of agents in a multi-agent system. The general idea is that agent  $i$  knows  $\varphi$  iff  $\varphi$  holds in all situations that  $i$  considers possible. The *de facto* standard framework for KBPs in multi-agent systems is the theory of systems and runs as described in the book *Reasoning About Knowledge* by Fagin, Halpern, Moses and Vardi ([8]). In this framework, the meaning (semantics) of a KBP is formulated in terms of the meaning of joint protocols, which are composed from individual (nondeterministic but sequential) protocols via synchronous parallelism.

The purpose of this paper is twofold: we want to use asynchronous parallelism as the basis for KBPs, and we intend to resolve the inherent circularity in the definition of the meaning of KBPs: on

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the one hand, the meaning of a KBP depends on the meaning of the knowledge operators for the agents involved, on the other hand, the meaning of the knowledge operators depends on the collection of possible executions of the KBP, i.e. on the meaning of the KBP. In [8], this circularity is resolved by giving an implicit definition of the meaning of a KBP (see Section 3 below). As a consequence, several KBPs have a unique meaning, some have no meaning at all, and others have more than one meaning. Our approach is to observe that the implicit definition of the meaning of a KBP can be read as a fixed point of an automorphism on the collection of possible meanings, and to define the *unique* meaning of the KBP in question as either the greatest fixed point of the automorphism if it exists, or a well-chosen iteration of the automorphism.

In our investigation, we bring together three traditions, that of knowledge-based programming, e.g. [8], that of semantics of sequential programming languages, e.g. [3, 14], and the semantics of concurrent programming, e.g. [1, 2].

Our reasons for this investigation are the following. Firstly, in [11] we were inspired by a problem from delay-insensitive circuits, which are inherently asynchronous. Asynchronous parallelism is, in general, more realistic in the context of multi-agent systems than synchronous parallelism. It is natural to assume that agents have different processing speeds. Asynchronicity was therefore our primary design goal. In combination with interleaving semantics, where the atomic actions are serialized in arbitrary order, it is the usual framework for concurrent programming [1, 2]. This has the technical advantage that the designer need not consider joint actions. The stutterings of [1] are introduced in our framework to model that agents cannot know the computation speeds of other agents.

Secondly, we are interested in the semantics of KBPs and proof rules for KBPs. For program design it is desirable to work with high-level programs as specifications. It is then useful that these programs have a unique meaning, so one can reason about the system under design, e.g. using proof rules based on this unique meaning. Since we aim at design by specification, we do not want to define the meanings of KBPs by means of standard protocols as in [8]. Therefore, we adapt the framework of Fagin et al. in such a way that we can eliminate the intermediate protocols.

Our investigation results in a number of semantic dilemmas. We therefore refrain from adding the complications of temporal operators and fairness. Consequently, we can restrict to finite prefixes of runs. So, instead of runs we use finite sequences of states, called traces. We prefer to investigate the best-possible semantics thoroughly, before making simplifications that would enable model checking.

## 1.1. Overview

In Section 2 we discuss related work. The general approach of Fagin et al. is outlined in Section 3. We adapt this general framework to suit our needs in Section 4. We define a language of KBPs, assign an interpretation to this language and propose how to assign semantics to all KBPs. Our choice of semantics is justified in Section 5 via a number of examples. Some conclusions and directions for future research are discussed in Section 6.

## 2. Several approaches to knowledge-based programming

The first papers on knowledge in distributed systems appear in the '80s: Chandy and Misra [5] describe how processes gain and lose knowledge, Katz and Taubenfeld [17] define various notions of knowledge,

depending on the state of information a process has. However, the programs in these papers are not KBPs in our sense, since they do not contain explicit tests for knowledge.

In an attempt to reason formally about KBPs and to gain more insight in KB-programming, Sanders [20] defines knowledge of a process using predicate transformers, and points out that safety and liveness properties of KBPs need not be preserved when the initial conditions are strengthened. Halpern and Zuck [12] successfully use the knowledge-based approach to derive and prove the correctness of a family of protocols for the sequence transmission problem. Their motivation for using a knowledge-based approach is that correctness proofs should also offer an understanding to the reader why a protocol is correct. Stulp and Verbrugge [21] show that real-life protocols can be analyzed using a knowledge-based approach by presenting a KBP for the TCP-protocol.

Moses and Kislev [19] introduce the notion of knowledge-oriented programs, i.e. KBPs with high-level actions that change the epistemic state of the agents. An example: the action  $notify(j, \varphi)$  ensures that agent  $j$  will eventually know  $\varphi$ . Formal semantics are not given, however. This kind of programs is also investigated more formally from the perspective of dynamic epistemic logic, e.g. by Baltag [4] and Van Ditmarsch [6]. The implementation of knowledge-oriented programs is not considered.

The book *Reasoning about Knowledge* [8], by Fagin et al., contains an overview of the work done on KB-programming by the authors and others. A general framework for KBPs is given, and the semantics of a KBP is defined via a standard (i.e. knowledge-free) program that implements the KBP. This knowledge-free program can be decomposed into a protocol for each agent that maps local states to actions. We briefly review this general framework in the next section. In [8, 9], sufficient conditions for the existence of a well-defined implementation of a KBP are given; these conditions only apply to synchronous systems, however. In [9] the complexity of determining whether a KBP has a well-defined implementation in a given finite state context is characterized. In [22], Vardi investigates the complexity of checking whether a protocol implements a KBP in a given finite state context.

Van der Meyden describes in [18] an axiomatization of a logic of knowledge and time for a class of synchronous and asynchronous systems and studies the model checking problem for this setting. However, the gap between writing down a KBP and determining its model remains. Engelhardt, van der Meyden and Moses develop in [7] a refinement calculus for KBPs. Their framework assumes that agents and environment act synchronously.

### 3. The general framework of Fagin, Halpern, Moses and Vardi

In this section we give a concise presentation of the relevant material from chapters 4, 5 and 7 of the book *Reasoning about Knowledge* [8], by Fagin et al.

#### 3.1. Global states, local states, runs and interpreted systems

A multi-agent system consists of an number of agents  $A = \{1, \dots, n\}$  in an environment. At each moment in time the system is in a certain *global state*  $\in \mathcal{G}$ :

$$\mathcal{G} = L_e \times L_1 \times \dots \times L_n$$

where for  $i \in A$ ,  $L_i$  is the set of *local states* of agent  $i$ , and  $L_e$  is the set of local states of the environment. If  $s \in \mathcal{G}$  and  $i \in A$  then  $s_i$  is the  $i + 1$ -th component of  $s$ , the local state of agent  $i$ .

A *run* captures how the global state changes over time. So, a run  $r$  is a function from  $\mathbb{N}$  to  $\mathcal{G}$  (time is taken to range over the natural numbers). A *point*  $(r, m)$  is a run  $r$  together with a time  $m$ , and the corresponding global state is  $r(m)$ . A *system*  $\mathcal{R}$  is a nonempty set of runs.

An *interpreted system*  $\mathcal{I}$  is a system  $\mathcal{R}$  together with an *interpretation function*  $\pi : \mathcal{G} \rightarrow (\Phi \rightarrow \{\text{true}, \text{false}\})$ , that assigns a valuation to a set  $\Phi$  of primitive propositions in each global state. The set  $\Phi$  may be partitioned into sets of local primitive propositions, i.e.  $\Phi = \cup_{i \in A} \Phi_i$ . An interpretation  $\pi$  is *compatible* if  $\pi$  is generated from local interpretations  $\pi_i : L_i \rightarrow \Phi_i \rightarrow \{\text{true}, \text{false}\}$ . So, for compatible  $\pi$ , we have that

$$\pi(s)(p) = \pi_i(s_i)(p) \text{ if } p \in \Phi_i. \quad (1)$$

**Definition.** Given a set  $\Phi$  of primitive propositions, the language  $\mathcal{L}(\Phi)$  of *epistemic formulas* is defined in BNF-notation as

$$\varphi ::= \perp \mid p \mid \neg\varphi \mid (\varphi \wedge \varphi) \mid K_i\varphi \quad (2)$$

where  $p \in \Phi$ ,  $i \in A$ . The connectives  $\vee$ ,  $\rightarrow$ ,  $\leftrightarrow$ , and the constant  $\top$  are defined as usual in terms of  $\perp$ ,  $\neg$  and  $\wedge$ .  $K_i$  is the knowledge operator for agent  $i$ . The possibility operator  $M_i$  is the dual of  $K_i$ :  $M_i\varphi = \neg K_i\neg\varphi$ .

An interpreted system  $\mathcal{I} = (\mathcal{R}, \pi)$  induces a *Kripke structure*  $M_{\mathcal{I}} = \langle S, \pi, \sim_1, \dots, \sim_n \rangle$  in the following way.  $S$  is the set of all points occurring in the runs of  $\mathcal{R}$ . The *indistinguishability relation*  $\sim_i$  for agent  $i$  is defined on points as

$$(r, m) \sim_i (r', m') \Leftrightarrow (r(m))_i = (r'(m'))_i$$

That is, agent  $i$  cannot distinguish between two global states  $s$  and  $s'$  if she has the same local state in both  $s$  and  $s'$ .

The *modeling relation*  $\models$  for an interpreted system  $\mathcal{I}$  is now taken to be the modeling relation of the associated Kripke model  $M_{\mathcal{I}}$ . More formally, if  $\mathcal{I}$  is an interpreted system and  $(r, m)$  is a point of  $\mathcal{I}$ , then

$$(\mathcal{I}, r, m) \models \varphi \Leftrightarrow (M_{\mathcal{I}}, r(m)) \models \varphi$$

for  $\varphi$  in  $\mathcal{L}(\Phi)$ . That is

$$\begin{aligned} (\mathcal{I}, r, m) \models p &\Leftrightarrow \pi(r(m))(p) = \text{true}, \text{ for } p \in \Phi \\ (\mathcal{I}, r, m) \models K_i\varphi &\Leftrightarrow \forall r' \in \mathcal{R} \forall m' (r'(m') \sim_i r(m) \Rightarrow (\mathcal{I}, r', m') \models \varphi) \end{aligned}$$

For the temporal operators  $\square$  (always) and  $\bigcirc$  (next), we have

$$\begin{aligned} (\mathcal{I}, r, m) \models \square\varphi &\Leftrightarrow \forall m' \geq m ((\mathcal{I}, r, m') \models \varphi) \\ (\mathcal{I}, r, m) \models \bigcirc\varphi &\Leftrightarrow (\mathcal{I}, r, m+1) \models \varphi \end{aligned}$$

but we shall not use  $\square$  and  $\bigcirc$  in the rest of this paper.

If  $\varphi$  is an atemporal, epistemic formula (i.e. contains no temporal operators) then  $(\mathcal{I}, r, m) \models \varphi$  only depends on  $\mathcal{I}$  and the global state  $s = r(m)$ , and we may write  $(\mathcal{I}, s) \models \varphi$  in that case. Moreover, if  $\varphi$  is a propositional formula (i.e. contains neither epistemic nor temporal operators) then  $(\mathcal{I}, s) \models \varphi$  only depends on the interpretation  $\pi$  instead of the entire interpreted system  $\mathcal{I}$ , and we may write  $(\pi, s) \models \varphi$  in that case. Finally, if  $\varphi$  is a propositional formula over  $\Phi_i$  and  $\pi$  is a compatible interpretation, then  $(\pi, s) \models \varphi$  only depends on agent  $i$ 's local state and so we may write  $(\pi, s_i) \models \varphi$  in that case.

### 3.2. Protocols and standard programs

For each agent  $i$ , a set  $ACT_i$  of *actions* available for that agent is given; the environment can execute actions from the set  $ACT_e$ . It is assumed that each set of actions contains a null-action,  $\Lambda$ , which has no effect. A *joint action* is tuple  $(a_e, a_1, \dots, a_n)$  of actions of the environment and the agents.  $ACT$  is the set of joint actions. There is a *transition function*  $\tau : ACT \rightarrow \mathcal{G} \rightarrow \mathcal{G}$  that indicates how joint actions may change the global state.

A *protocol* is a function from local states to nonempty sets of actions, i.e. a protocol for agent  $i$  is a function  $P_i : L_i \rightarrow (\mathcal{P}(ACT_i) - \{\emptyset\})$ . A *joint protocol* is a tuple  $(P_1, \dots, P_n)$  of protocols for each of the agents.

A *context*  $\gamma$  is a tuple  $(P_e, \tau, \mathcal{G}_0, \Psi)$ , consisting of the protocol  $P_e$  of the environment, the transition function  $\tau$ , a set of initial global states  $\mathcal{G}_0 \subseteq \mathcal{G}$ , and an admissibility restriction  $\Psi \subseteq (\mathbb{N} \rightarrow \mathcal{G})$  on runs.

A run  $r$  is *consistent* with joint protocol  $P = (P_1, \dots, P_n)$  in context  $\gamma = (P_e, \tau, \mathcal{G}_0, \Psi)$  if

- $r(0) \in \mathcal{G}_0$ ,
- $\forall m \in \mathbb{N} \exists \vec{a} \in \vec{P}(r(m)) \exists a_e \in P_e(r_e(m)) (r(m+1) = \tau(a_e, \vec{a})(r(m)))$ ,
- $r \in \Psi$ .

where  $\vec{P}(s)$  is the result of applying the joint protocol to a global state:

$$\vec{P}(s_e, s_1, \dots, s_n) = P_1(s_1) \times \dots \times P_n(s_n).$$

Observe that the second formula in the definition of consistency implies that the joint protocol is executed not once, but infinitely often.

The *representing system*  $\mathcal{R}^{\text{rep}}(P, \gamma)$  of joint protocol  $P$  in context  $\gamma$  is defined by

$$\mathcal{R}^{\text{rep}}(P, \gamma) = \{r \in \mathbb{N} \rightarrow \mathcal{G} \mid r \text{ is consistent with } P \text{ in } \gamma\}$$

Protocols are represented by programs. A *standard program*  $Pg_i$  for agent  $i$  has the following form

```

case of
  if  $t_1$  do  $a_1$ 
   $\vdots$ 
  if  $t_m$  do  $a_m$ 
end case

```

where the  $t_j$  are propositional formulas over  $\Phi_i$  and the  $a_j$  are in  $ACT_i$ .

Given a program  $Pg_i$  and a compatible interpretation  $\pi$ , the *associated protocol*  $Pg_i^\pi$  in local state  $l \in L_i$  is defined as

$$Pg_i^\pi(l) = \begin{cases} \{a_j \mid (\pi, l) \models t_j\} & \text{if } \{j \mid (\pi, l) \models t_j\} \neq \emptyset \\ \{\Lambda\} & \text{otherwise} \end{cases}$$

The compatibility of  $\pi$  (see (1)) ensures that the outcome of tests  $t_j$  in program  $Pg_i$  under  $\pi$  only depends on the local state of agent  $i$ , so we may write  $(\pi, l) \models t_j$  for  $(\mathcal{I}, r, m) \models t_j$ . So the associated protocol

chooses nondeterministically one of the actions  $a_j$  for which the test  $t_j$  holds; if there is no such  $a_j$  then the empty action  $\Lambda$  is chosen.

A *joint program* is a tuple  $Pg = (Pg_1, \dots, Pg_n)$ . Given an interpretation  $\pi$ , compatible with every program of the joint program, the joint protocol  $Pg^\pi$  denoted by  $Pg$  is defined straightforwardly.

An *interpreted context* is a context  $\gamma$  with an interpretation  $\pi$ , that is a tuple  $(\gamma, \pi)$ . The interpreted system of all runs that are consistent with  $P$  in interpreted context  $(\gamma, \pi)$  is denoted by

$$\mathcal{I}^{\text{rep}}(P, \gamma, \pi) = (\mathcal{R}^{\text{rep}}(P, \gamma), \pi).$$

This is the interpreted system that represents  $P$  in interpreted context  $(\gamma, \pi)$ .

Observe that the infinite repetition of joint program  $Pg$ , which is a consequence of the definition of their interpretation in terms of consistent runs, is left implicit in the notation. The same holds for the knowledge-based programs defined in [8]: see the next section. We shall make this repetition explicit in our adapted definition of knowledge-based programs in Section 4.4.

### 3.3. Knowledge-based programs

A *knowledge-based program* (KBP) for agent  $i$  is of the form

**case of**  
     **if**  $t_1 \wedge k_1$  **do**  $a_1$   
     :  
     **if**  $t_m \wedge k_m$  **do**  $a_m$   
**end case**

where the  $t_j$  are propositional formulas over  $\Phi_i$ , the  $a_j$  are in  $ACT_i$  and the  $k_j \in \mathcal{L}(\Phi)$  are epistemic formulas as defined in (2). A *joint KBP* is a tuple of KBPs.

In contrast to the propositional formulas  $t_j$ , the outcome of epistemic formulas  $k_j$  does not only depend on the interpretation  $\pi$  and the local state of the agent, but on the entire interpreted system. So the joint protocol  $Pg^\mathcal{I} = (Pg_1^\mathcal{I}, \dots, Pg_n^\mathcal{I})$  has  $\mathcal{I}$  instead of  $\pi$  as a parameter. As before, the interpretation  $\pi$  in  $\mathcal{I}$  should be compatible with  $Pg$  w.r.t. the propositional formulas  $t_j$ . There is no restriction on the propositions in the epistemic formulas  $k_j$ .

The modeling relation is extended to evaluate epistemic formulas for agent  $i$  in a local state  $l \in L_i$  for a compatible interpretation  $\pi$ . If  $\varphi$  is a propositional formula, then  $(\mathcal{I}, l) \models \varphi$  iff  $(\pi, l) \models \varphi$ . For epistemic formulas,

$$(\mathcal{I}, l) \models K_i \psi \quad \Leftrightarrow \quad \forall r, m \ ( (r, m) \in \mathcal{I} \wedge (r(m))_i = l \Rightarrow (\mathcal{I}, r, m) \models \psi )$$

A protocol  $Pg_i^\mathcal{I}$  is associated with KBP  $Pg_i$  for  $l \in L_i$  as

$$Pg_i^\mathcal{I}(l) = \begin{cases} \{a_j \mid (\mathcal{I}, l) \models t_j \wedge k_j\} & \text{if } \{j \mid (\mathcal{I}, l) \models t_j \wedge k_j\} \neq \emptyset \\ \{\Lambda\} & \text{otherwise } \emptyset \end{cases}$$

An interpreted system  $\mathcal{I} = (\mathcal{R}, \pi)$  represents a KBP  $Pg$  in context  $(\gamma, \pi)$  iff  $\pi$  is compatible with  $Pg$  and  $\mathcal{I}$  represents the associated protocol  $Pg^\mathcal{I}$ . That is,  $\mathcal{I}$  represents  $Pg$  in  $(\gamma, \pi)$  iff

$$\mathcal{I} = \mathcal{I}^{\text{rep}}(Pg^\mathcal{I}, \gamma, \pi) \tag{3}$$

Observe that we have a circularity here: to determine whether an interpreted system  $\mathcal{I}$  represents  $Pg$  we must verify whether the runs of  $\mathcal{I}$  are consistent with  $Pg^{\mathcal{I}}$ . So this is an implicit definition, and there need not be a unique interpreted system that represents a KBP in a given context.

This ends our resumé of the theory of systems and runs as given in [8].

## 4. Asynchronous KBPs

For our treatment of asynchronous KBPs, we shall adapt the framework of systems and runs given in the previous section. The main changes are the following.

- We combine actions via interleaving with possible delay (and not via joint actions).
- We work with a more general programming language, inspired by dynamic logic (see [13]) which is interpreted directly (i.e. not via protocols) in a Kripke model.
- The indistinguishability relations  $\approx_i$  in our Kripke model are relations on runs (not on states), so we assume that the agents can recall their own history when comparing points in the model.
- In the definition of the interpretation of programs in our Kripke model, we introduce an explicit parameter  $B$  to restrict the collection of runs under consideration when interpreting the knowledge operators  $K_i$ ; via abstraction from  $B$ , we obtain an automorphism on our Kripke model which is the starting point for the definition of the (unique and parameter-free) meaning of our programs.

Some minor changes are: we consider the environment as an agent like the others, we shall shift from infinite runs to finite traces (both approaches are equivalent when restricted to atemporal formulas), and we work with a slightly different notion of context.

### 4.1. Asynchronicity via stutterings

To model asynchronicity, we work with an interleaving semantics: the single actions of the individual programs of the agents are performed consecutively in an unspecified order, possibly with delay. As a consequence, an agent cannot distinguish two runs  $r, r' \in (\mathbb{N} \rightarrow \mathcal{G})$  if  $r'$  is obtained from  $r$  by consecutive repetition of certain global states  $s \in G$ .

We illustrate this with an example. Assume that agent  $a$  can see the contents of numerical variable  $x$ , and that agents  $b$  and  $c$  can modify  $x$ :  $b$  can increase it with 2,  $c$  can decrease it with 1. In our setting, 02133333... is a possible run, where every step corresponds with either an action of  $b$ , an action of  $c$  or no action; 01355555... is not possible, for the first step (going from 0 to 1) requires combining the actions of  $b$  and  $c$  in one joint action, which is not allowed in interleaving semantics. Moreover, the runs  $r = 021354333...$  and  $r' = 000213333555443333...$  cannot be distinguished by agent  $a$ , for both correspond with actions from  $b, c, b, b, c, c$ , respectively, and we abstract away from the delay between some of the actions that occurred in  $r'$ .

**Definition.** A run  $r_2$  is a *stuttering* of  $r_1$ , notation  $r_1 \preceq r_2$ , iff  $r_2$  is obtained from  $r_1$  by consecutive, finite repetition of certain elements of  $r_1$ . In formula:

$$r_1 \preceq r_2 \quad \equiv \quad \exists f : \mathbb{N} \rightarrow \mathbb{N} \text{ (} f \text{ monotonic and surjective, and } r_2 = r_1 \circ f \text{)} \quad (4)$$



For example,  $r'$  of the example above is a stuttering of  $r$ , but  $0211111\dots$  is not (infinite repetition of 1), nor  $012354333\dots$  (the order of the elements is not preserved). The concept of stuttering comes from [1], and the relation  $\preceq$  was introduced in [15]. We claim that  $\preceq$  is a partial order: reflexivity and transitivity follow directly from the definition, antisymmetry is a nice nontrivial exercise.

With this notion of stuttering, we can define the appropriate indistinguishability relation for agents in the context of asynchronous KBPs.

**Definition.** Agent  $i$  considers run  $r_1$  indistinguishable from run  $r_2$  (notation:  $r_1 \approx_i r_2$ ) when both runs can be stuttered to runs  $r'_1$  and  $r'_2$  such that agent  $i$  cannot distinguish individual states along  $r'_1$  and  $r'_2$ . So we have

$$r_1 \approx_i r_2 \quad \equiv \quad \exists r'_1, r'_2 (r_1 \preceq r'_1 \wedge r_2 \preceq r'_2 \wedge r'_1 \sim_i r'_2). \quad (5)$$

In short:  $(\approx_i) = (\preceq \circ \sim_i \circ \succeq)$ , where  $\circ$  denotes relational composition. Here we use the straightforward lifting of  $\sim_i$  to runs, defined by

$$r_1 \sim_i r_2 \quad \equiv \quad \forall n (r_1(n) \sim_i r_2(n)) .$$

We show that  $\approx_i$  is an equivalence relation. It is easy to see that  $\approx_i$  is reflexive and symmetrical, since  $\sim_i$  is reflexive and symmetrical and  $\preceq$  is reflexive. For transitivity, i.e.  $(\approx_i \circ \approx_i) \subseteq (\approx_i)$ , we use the confluence of  $\preceq$  and the fact that  $(\sim_i \circ \preceq) \subseteq (\preceq \circ \sim_i)$ :

$$\begin{aligned} & \approx_i \circ \approx_i \\ = & \quad \{\text{definition}\} \\ & \preceq \circ \sim_i \circ \succeq \circ \preceq \circ \sim_i \circ \succeq \\ \subseteq & \quad \{\text{confluence of } \preceq: (\succeq \circ \preceq) \subseteq (\preceq \circ \succeq)\} \\ & \preceq \circ \sim_i \circ \preceq \circ \succeq \circ \sim_i \circ \succeq \\ \subseteq & \quad \{(\sim_i \circ \preceq) \subseteq (\preceq \circ \sim_i) \text{ and } (\succeq \circ \sim_i) \subseteq (\sim_i \circ \succeq)\} \\ & \preceq \circ \preceq \circ \sim_i \circ \sim_i \circ \succeq \circ \succeq \\ = & \quad \{\text{transitivity of } \sim_i \text{ and } \preceq\} \\ & \preceq \circ \sim_i \circ \succeq \\ = & \quad \{\text{definition}\} \\ & \approx_i \end{aligned}$$

## 4.2. A Kripke model of traces

In this paper, we study only KBPs with purely epistemic tests (the extension to tests with temporal operators is an interesting subject for further research). In the last paragraph of Subsection 3.1, we observed that in that case we may restrict ourselves, without loss of generality, to finite nonempty prefixes of runs, i.e.  $\text{traces } xs \in \mathcal{G}^+$ . We shall do so from now on, and introduce some notation. We write  $\ell(xs)$  for the length of the trace, and  $xs_j$  for its  $j$ th element (where  $0 \leq j < \ell(xs)$ ). The function  $\text{last}$  returns the last element of a trace, i.e.  $\text{last}(xs) = xs_{n-1}$  where  $\ell(xs) = n$ . Concatenation of traces is denoted by  $*$ . The definition of stuttering and indistinguishability for traces is analogous to that of runs.

We intend to define the semantics of KBPs, given a context  $\gamma = (\mathcal{G}, \pi, \tau, \mathcal{G}_0)$ . As before,  $\mathcal{G}$  is the collection of global states,  $\pi : \mathcal{G} \rightarrow (\Phi \rightarrow \{\text{true}, \text{false}\})$  is a valuation of atomic propositions,  $\tau : ACT \rightarrow \mathcal{G} \rightarrow \mathcal{G}$  is a transition function for the interpretation of atomic actions and  $\mathcal{G}_0 \subseteq \mathcal{G}$  is a collection of initial states.

We now present the Kripke model  $M = \langle \mathcal{G}^+, \pi, (\approx_i)_{i \in A} \rangle$  for the interpretation of knowledge formulas and knowledge-based programs. The set of worlds is the collection  $\mathcal{G}^+$  of all traces over  $\mathcal{G}$ .  $\pi$  is a valuation given by the context, and  $(\approx_i)_{i \in A}$  is the collection of equivalence relations on traces defined above.

The interpretation of epistemic formulas in  $M$  is defined as follows. We use a parameter  $B$ , denoting the collection of traces *under consideration*, to resolve the mutual dependency between the definition of the semantics of the knowledge operator and the definition of the collection of possible traces. For  $B \subseteq \mathcal{G}^+$ , the interpretation  $\llbracket \varphi \rrbracket_B \subseteq \mathcal{G}^+$  is defined inductively by

$$\begin{aligned} \llbracket \perp \rrbracket_B &= \emptyset, \\ \llbracket p \rrbracket_B &= \{xs \in \mathcal{G}^+ \mid \pi(\text{last}(xs))(p) = \text{true}\}, \\ \llbracket \varphi_1 \wedge \varphi_2 \rrbracket_B &= \llbracket \varphi_1 \rrbracket_B \cap \llbracket \varphi_2 \rrbracket_B, \\ \llbracket \neg \varphi \rrbracket_B &= \mathcal{G}^+ \setminus \llbracket \varphi \rrbracket_B, \\ \llbracket K_i \varphi \rrbracket_B &= \{xs \in \mathcal{G}^+ \mid \forall ys \in B (xs \approx_i ys \Rightarrow ys \in \llbracket \varphi \rrbracket_B)\}. \end{aligned}$$

So a trace  $xs$  satisfies proposition  $p$  iff  $p$  holds in the last state of  $xs$  under valuation  $\pi$ , and agent  $i$  knows that  $\varphi$  holds in trace  $xs$  iff  $\varphi$  holds in all traces  $ys$  under consideration that  $i$  cannot distinguish from  $xs$ .

Observe that the role played by  $B$  differs from the more global role of  $\mathcal{R}$  in the definition of the interpretation of formulas in interpreted system  $\mathcal{I} = (\mathcal{R}, \pi)$  used in [8] (see Section 3). Here, we work in the Kripke model of *all* possible traces, and  $B$  is only used as a restriction in the definition of the interpretation of  $K_i$ ; as a consequence, we do not have  $\llbracket \varphi \rrbracket_B \subseteq B$  in general. This is a deliberate choice to be justified later by means of the example programs in Section 5.

### 4.3. Epistemic programs

**Definition.** Given a set  $ACT$  of atomic actions  $a$ , the language of *epistemic programs* is defined as

$$\alpha ::= \Lambda \mid a \mid \varphi? \mid (\alpha; \alpha) \mid (\alpha \cup \alpha) \mid \alpha^* \quad (6)$$

where  $a \in ACT$ ,  $\varphi$  an epistemic formula. So a program is either a null action  $\Lambda$ , an atomic action  $a$ , a test on  $\varphi$ , the composition of two programs, the nondeterministic choice between two programs, or the repetition (zero or more times) of a program.

The interpretation of an epistemic program is a binary relation on traces. As for formulas, we have a parameter  $B \subseteq \mathcal{G}^+$  that represents the traces under consideration. Now the definition of  $\llbracket \alpha \rrbracket_B \subseteq \mathcal{G}^+ \times \mathcal{G}^+$  reads

$$\begin{aligned} \llbracket \Lambda \rrbracket_B &= \{(xs, xs * \text{last}(xs)) \mid xs \in \mathcal{G}^+\} \\ \llbracket a \rrbracket_B &= \{(xs, xs * z \mid xs \in \mathcal{G}^+ \wedge z = \tau(a)(\text{last}(xs))\} \\ \llbracket \varphi? \rrbracket_B &= \{(xs, xs) \mid xs \in \llbracket \varphi \rrbracket_B\}, \\ \llbracket \alpha_1 ; \alpha_2 \rrbracket_B &= \llbracket \alpha_1 \rrbracket_B \circ \llbracket \alpha_2 \rrbracket_B, \\ \llbracket \alpha_1 \cup \alpha_2 \rrbracket_B &= \llbracket \alpha_1 \rrbracket_B \cup \llbracket \alpha_2 \rrbracket_B, \\ \llbracket \alpha^* \rrbracket_B &= \bigcup_{n \geq 0} (\llbracket \alpha \rrbracket_B)^n \end{aligned}$$

We give some explanation.  $\Lambda$  (doing the empty action) corresponds with a stuttering step: the last (i.e. current) state is repeated once. Observe that  $\Lambda$  is different from  $\top?$  (not acting), since  $\llbracket \top? \rrbracket_B =$

$\{(xs, xs) \mid xs \in \mathcal{G}^+\}$ . The interpretation of atomic actions is lifted to traces: two traces  $xs$  and  $ys$  are related via action  $a$  iff  $ys$  is equal to  $xs$  appended with an outcome of  $a$  applied to the last element in  $xs$ . The interpretation of the program constructs (composition, choice and repetition) follows the usual treatment in dynamic logic.

#### 4.4. Knowledge-based programs

Recall the definition of knowledge-based program  $Pg_i$  of agent  $i \in A = \{1, \dots, n\}$ , given in Section 3.3. In our formalism, it can be rendered as the repetition of a nondeterministic choice of guarded atomic actions:

$$Pg_i = \left( \bigcup_{j \in J_i} (\varphi_j?; a_j) \right)^*$$

where  $J_i$  is some finite index set for  $i \leq n$ . The implicit repetition of the KBPs of Section 3.3 is made explicit here by Kleene's star. In interleaving semantics, the parallel composition  $Pg$  of  $Pg_1, \dots, Pg_n$  is a nondeterministic repetition of all the alternatives  $\varphi_j?; a_j$  that occur in the different programs, with the alternative  $\Lambda$  added to model possible delay. So

$$Pg = \left( \Lambda \cup \bigcup_{j \in J} (\varphi_j?; a_j) \right)^* \quad (7)$$

where  $J = J_1 \cup \dots \cup J_n$ . In the rest of this paper, we restrict ourselves to programs of this form, which we call *asynchronous* KBPs. The interpretation  $\llbracket Pg \rrbracket_B$  can be rewritten as

$$\begin{aligned} \llbracket Pg \rrbracket_B = \{ & (xs * x_0, xs * x_0 * x_1 * \dots * x_n) \mid xs \in \mathcal{G}^* \wedge n \geq 0 \wedge \\ & \forall k < n (x_k = x_{k+1} \vee \exists j (xs * x_0 * x_1 * \dots * x_k \in \llbracket \varphi_j \rrbracket_B \wedge x_{k+1} = \tau(a_j)(x_k))) \} \end{aligned} \quad (8)$$

That is,  $\llbracket Pg \rrbracket_B$  consists of pairs of traces  $(ys, zs)$  where  $zs$  is an extension of  $ys$ , such that the new subsequent states are either the result of stuttering (i.e. doing the empty action  $\Lambda$ ), or of some atomic action  $a_j$  in  $Pg$  with a guard  $\varphi_j$  that holds in the trace up to that state.

Now the *set of traces generated by  $Pg$* , given a context  $\gamma = (\mathcal{G}, \pi, \tau, \mathcal{G}_0)$  and a set  $B$  of traces under consideration, is defined as the set of all traces that are generated by  $Pg$  when starting in an initial state  $y \in \mathcal{G}_0 \subseteq \mathcal{G}$ :

$$\mathcal{G}_0 \llbracket Pg \rrbracket_B = \{xs \mid \exists y (y \in \mathcal{G}_0 \wedge (y, xs) \in \llbracket Pg \rrbracket_B)\} \quad (9)$$

We would like to define the semantics of a program  $Pg$  to be some set of traces  $B$  such that  $Pg$  generates the set  $B$  when  $B$  is the set of traces under consideration, i.e.  $B$  should satisfy  $\mathcal{G}_0 \llbracket Pg \rrbracket_B = B$ . This is a fixpoint characterization of the semantics of program  $Pg$ . It is our analogue to equation (3) from Section 3.3.

#### 4.5. Choosing a semantics for KBPs

In this section we set out to find a unique, parameter-free semantics for asynchronous KBPs of the form given in (7). For that purpose, we define an automorphism  $F : \mathcal{P}(\mathcal{G}^+) \rightarrow \mathcal{P}(\mathcal{G}^+)$ , based on the definition of the trace set of  $Pg$  given in (9):

$$F(B) = \mathcal{G}_0 \llbracket Pg \rrbracket_B \quad (10)$$

If  $F$  is monotonic (in the sense that  $F(B) \subseteq F(B')$  whenever  $B \subseteq B'$ ), then the theorem of Knaster-Tarski tells us that  $F$  has a unique least fixpoint and a unique greatest fixpoint. In that case, we prefer to define the semantics of  $Pg$  as the *greatest* fixpoint, since that would be the most liberal interpretation of  $Pg$ . In general, however,  $F$  is not monotonic and may have no fixpoints at all, or it may have distinct fixpoints without having a least or greatest one. We will show this in the examples of Section 5.

Fixpoint equations are common in the study of the semantics of sequential programs with loops or recursive procedures (see e.g. [3, 14]), and the fixpoint is usually approximated by iterations of the fixpoint operator. We generalize that idea in a more general setting here, viz. when there may be no fixpoint to approximate.

**Definition.** The iterations  $(B_\lambda)$  are defined by transfinite induction over the ordinals:

$$\begin{aligned} B_0 &= \mathcal{G}^+, \\ B_{\lambda+1} &= F(B_\lambda), & \text{for any ordinal } \lambda, \\ B_\lambda &= \bigcap_{\mu < \lambda} \bigcup_{\mu \leq \nu < \lambda} B_\nu, & \text{for any limit ordinal } \lambda. \end{aligned}$$

where  $F$  is as defined in (10).

The first iteration  $B_0$  consists of all nonempty finite sequences of states. For any ordinal  $\lambda$ , the iteration  $B_{\lambda+1}$  consists of traces generated by the program when  $B_\lambda$  is used to evaluate epistemic formulas. When  $\lambda$  is a limit ordinal, the iteration  $B_\lambda$  is the intersection of unions of iterations that are sufficiently close to the limit. We motivate the definition of  $B_\lambda$  for a limit ordinal  $\lambda$  as follows. If the iteration sequence forms a descending chain, then for the limit an intersection is needed. On the other hand, if the sequence forms an ascending chain, then for the limit a union is needed. However, iterations may also grow and shrink, and therefore a union of intersections or an intersection of unions is indicated. Both result in an intersection for a descending sequence, and in a union for an ascending sequence. We choose the intersection of unions, as it is more liberal than a union of intersections: it includes more traces, more traces implies less knowledge and we want the agents to know facts only when there are good reasons for them.

A cardinality argument implies that the transfinite sequence of contains multiple elements: since all sets  $B_\lambda$  are subsets of  $\mathcal{G}^+$  and there exist more ordinals than  $\mathcal{G}^+$  has subsets, the sets  $B_\lambda$  cannot all be different. This implies the existence of ordinals  $\kappa$  and  $\mu$  with  $\kappa < \mu$  and  $B_\kappa = B_\mu$ . By well-foundedness, there is a least such  $\kappa$ . Now let  $\kappa$  be minimal with the property that  $B_\kappa = B_\mu$  for some  $\mu > \kappa$ . If  $F(B_\kappa) = B_\kappa$ , i.e.  $B_\kappa$  is a fixpoint, then we choose  $B_\kappa$  as the semantics for the program. Otherwise, we take the smallest  $\lambda \geq \kappa$  such that  $F(B_\lambda) \not\subseteq B_\lambda$  as the semantics for the program (such a  $\lambda$  always exist when  $B_\kappa$  is not a fixpoint). This latter choice is justified by the argument that it allows as much well-justified knowledge as possible without introducing contradictory knowledge.

**Definition.** We define

$$\text{sem}(Pg) = B_\lambda \tag{11}$$

where

$$\begin{aligned} \lambda &= \inf\{\lambda \mid \kappa \leq \lambda \wedge (B_\lambda = B_{\lambda+1} \vee B_{\lambda+1} \not\subseteq B_\lambda)\} \\ \kappa &= \inf\{\kappa \mid \exists \mu. (\kappa < \mu \wedge B_\kappa = B_\mu)\} \end{aligned}$$

Since the iteration sequence has multiple elements, it has the shape of a 6, see figure 1. All iterations between  $B_\kappa$  and  $B_\lambda$  are subsets of previous iterations, and  $B_\lambda$  is the last one for which this holds.

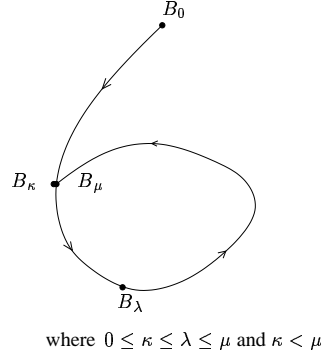


Figure 1. The iteration sequence.  $\kappa$  is the least ordinal such that  $\exists \mu > \kappa. (B_\kappa = B_\mu)$

When determining the traces in an iteration, it is sometimes useful to restrict attention to traces generated by the (knowledge-free) *flat program* obtained from  $Pg$  by omitting all guards:

$$Pg_b = (\Lambda \cup \bigcup_{j \in J} a_j)^*$$

We put  $B_b = \llbracket Pg_b \rrbracket$ ; observe that  $\llbracket Pg_b \rrbracket_B = \llbracket Pg_b \rrbracket_{B'}$  for all  $B, B' \subseteq \mathcal{G}^+$ , so we may drop the parameter  $B$  here.

## 5. Examples

In this section we investigate the consequences of our formalism by means of a number of examples. The first three examples serve to show that function  $F$  can be monotonic or not monotonic, and may have incomparable fixpoints or no fixpoints at all. The fourth example shows that the iteration sequence may become infinite. The fifth example demonstrates how message passing can be modelled. The final example is a formulation of the Unexpected Hanging Paradox as an asynchronous KBP. For every example the iteration sequence is also given as a figure.

### 5.1. Failure of monotonicity

Suppose there are two agents 1 and 2 and two boolean variables  $p, q$ , initially true. We represent truth values as the integers 0 and 1, so the state space is  $\mathcal{G} = \{0, 1\} \times \{0, 1\}$  with initial state  $(1, 1)$ . The first component of the state is the value of variable  $p$ , the second one is the value of  $q$ . This induces a valuation on states.

Variable  $p$  is private to agent 1, and  $q$  is private to agent 2. That is, agent 1 can only see and write  $p$ , agent 2 can only see and write  $q$ . We take the indistinguishability relations on states to be defined by

$$(p, q) \sim_1 (p', q') \Leftrightarrow p = p', \quad (p, q) \sim_2 (p', q') \Leftrightarrow q = q'.$$

Here and henceforth we use  $p, p', q, q'$  to denote the values of the program variables  $p$  and  $q$  in the states considered.

The agents execute KBPs. Agent 1 can falsify  $p$  by setting  $p$  to zero when she knows that  $q$  holds. Agent 2 can falsify  $q$  when she knows that  $p$  holds. We take assignments to variables as primitive action symbols, with corresponding interpretation. The resulting KBP is thus

$$Pg = (\Lambda \cup K_1q? ; p := 0 \cup K_2p? ; q := 0)^*.$$

By definition  $B_0 = \mathcal{G}^+$ . For the next iteration, the traces in  $B_0$  are used to evaluate epistemic tests. When  $\mathcal{G}^+$  is considered, agents have only knowledge of the values of their own private variables. That is, if  $B = \mathcal{G}^+$  then  $\llbracket K_1q \rrbracket_B = \emptyset$  and  $\llbracket K_2p \rrbracket_B = \emptyset$ . Both agents are unable to act. Therefore,  $B_1 = F(B_0) = (1, 1)^+$ . So, traces in  $B_1$  are repetitions of the initial state.

All traces in  $B_1$  satisfy both  $p$  and  $q$ . Therefore, if  $B = B_1$  then  $\llbracket K_1q \rrbracket_B = \llbracket K_2p \rrbracket_B = \mathcal{G}^+$ . This implies that the agents can independently set their private variable to zero. Therefore,  $B_2 = F(B_1)$  consists of traces generated by the knowledge-free program

$$(\Lambda \cup p := 0 \cup q := 0)^*.$$

Alternatively, we can say that  $B_2$  consists of traces that match the regular expression

$$(1, 1)^+ \mid (1, 1)^+(0, 1)^+(0, 0)^* \mid (1, 1)^+(1, 0)^+(0, 0)^*.$$

When the agents can independently set their private variable to zero, they have no knowledge on the value of the other agent's variable. That is, if  $B = B_2$  then  $\llbracket K_1q \rrbracket_B = \llbracket K_2p \rrbracket_B = \emptyset$ . Therefore  $B_3 = F(B_2) = (1, 1)^+$ , and so  $B_3 = B_1$  and the iteration sequence starts to repeat itself. It seems reasonable to take  $B_1$  as the correct meaning of the program. Since  $B_2$  is strictly larger than  $B_1$ , this is in accordance with the definition of the semantics given in (11).

The above iterations also show that  $F$  is not monotonic. For example,  $B_1$  is a subset of  $B_2$ , however,  $F(B_1)$  is not a subset of  $F(B_2)$ . Furthermore, in this example  $F$  has the following two unordered fixpoints:

$$\begin{aligned} B' &= (1, 1)^+(0, 1)^*, \\ B'' &= (1, 1)^+(1, 0)^*. \end{aligned}$$

To show that  $B'$  is a fixpoint of  $F$ , observe that all traces in  $B'$  satisfy  $q$ , so  $\llbracket K_1q \rrbracket_{B'} = \mathcal{G}^+$ . Therefore, agent 1 can set  $p$  to zero at any time. Traces matching  $(1, 1)^+$  or  $(1, 1)^+(0, 1)^+$  are both indistinguishable for agent 2 from traces of the form  $(1, 1)^+(0, 1)^+$ , and therefore do not satisfy  $K_2p$  under  $B'$ . Agent 2 cannot set  $q$  to zero in the initial state nor after agent 1 has reset  $p$ , so  $F(B') = B'$ . Analogously,  $F(B'') = B''$ .

The iteration sequence and the fixpoints  $B'$  and  $B''$  are sketched in figure 2. The dashed line indicates the subset relation: the lower set is a subset of the upper set. Arrows give the direction of iteration, i.e. application of  $F$ .

## 5.2. A case with fixpoint semantics

This example has the same setting as the previous example. There are two agents 1 and 2, with private boolean variables  $p$  respectively  $q$ . The state space is as before, but now the initial state is  $(0, 0)$ . An

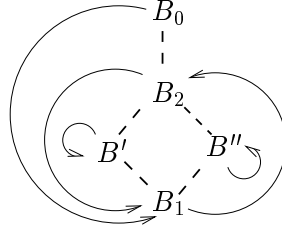
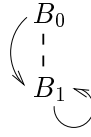
Figure 2. Example 5.1: iteration sequence and the fixpoints  $B'$  and  $B''$ 

Figure 3. Example 5.2: iteration sequence that ends in a fixpoint

agent may set her private variable to 1 if she considers it possible that the other agent's private variable has already been set. This corresponds to the following program

$$Pg = (\Lambda \cup M_1q? ; p := 1 \cup M_2p? ; q := 1)^*.$$

Again we determine a sequence of iterations. When  $B_0 = \mathcal{G}^+$  is considered, agents have only knowledge of their private variables. If agent 1 does not know whether  $q$  holds then she considers both  $q$  and  $\neg q$  possible. So, if  $B = \mathcal{G}^+$  then  $\llbracket M_1q \rrbracket_B = \llbracket M_2p \rrbracket_B = \mathcal{G}^+$ . The agents can independently set their private variables, so  $B_1 = F(B_0)$  consists of the traces generated by

$$(\Lambda \cup p := 1 \cup q := 1)^*.$$

In  $B_1$ , agents always consider it possible that the other has set her variable. Therefore, if  $B = B_1$  then  $\llbracket M_1q \rrbracket_B = \llbracket M_2p \rrbracket_B = \mathcal{G}^+$ , so  $B_2 = F(B_1) = B_1$ , and we have reached a fixpoint; as such, it is the semantics of the program according to the definition of the semantics given in (11). (Note that  $(0, 0)^+$  is another fixpoint of  $F$ .) The iteration is sketched in figure 3.

Anthropomorphically speaking, we see that either agent reckons with the possibility that the other agent has set her variable, even though she can argue that the other agent should not be able to do so as the first one. This is an unexpected side effect of the asynchrony.

### 5.3. Absence of fixpoints

Again there are two agents 1 and 2 with private boolean variables  $p$  and  $q$ . Initially  $p$  is false and  $q$  is true. Now, agent 1 can set  $p$  when she knows that  $q$  holds, and agent 2 can reset  $q$  when she considers it possible that  $p$  has been set. The program is given as

$$Pg = (\Lambda \cup K_1q? ; p := 1 \cup M_2p? ; q := 0)^*.$$

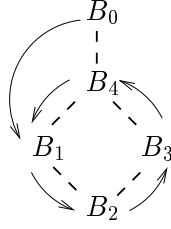


Figure 4. Example 5.3: a concrete six-shaped diagram as given in figure 1

As seen in the previous two examples, if  $B = \mathcal{G}^+$  then  $\llbracket K_1 q \rrbracket_B = \emptyset$  and  $\llbracket M_2 p \rrbracket_B = \mathcal{G}^+$ . Only agent 2 can act, so  $B_1 = F(B_0)$  consists of traces of program

$$(\Lambda \cup q := 0)^*.$$

That is,  $B_1$  contains traces that match  $(0, 1)^+(0, 0)^*$ .

For  $B = B_1$ , the interpretation  $\llbracket K_1 q \rrbracket_B$  contains traces that cannot be generated by the program. For example, a trace  $xs$  matching  $(1, 0)^+$  satisfies  $K_1 q$  under  $B_1$ , since no trace in  $B_1$  is indistinguishable for agent 1 from  $xs$ . Therefore, we determine the set  $B_\flat$  of all traces that can possibly be generated by program  $Pg$ . The program only allows  $p$  to change from 0 to 1, and  $q$  from 1 to 0. So  $B_\flat$  is the set of all traces of the program in which the guards have been removed, that is the program

$$(\Lambda \cup p := 1 \cup q := 0)^*.$$

No trace in  $B_\flat$  satisfies  $K_1 q$  under  $B_1$ . In other words, if  $B = B_1$  then  $B_\flat \cap \llbracket K_1 q \rrbracket_B = \emptyset$ . No trace in  $B_1$  satisfies  $p$ . Therefore, if  $B = B_1$  then  $\llbracket M_2 p \rrbracket_B = \emptyset$ . Both agents can never act, so  $B_2 = F(B_1) = (0, 1)^+$ .

All traces in  $B_2$  satisfy  $\neg p$  and  $q$ , so for  $B = B_2$  we have  $\llbracket K_1 q \rrbracket_B = \mathcal{G}^+$  and  $\llbracket M_2 p \rrbracket_B = \emptyset$ . Only agent 1 can act. Therefore,  $B_3 = F(B_2)$  consists of traces generated by program

$$(\Lambda \cup p := 1)^*.$$

Since  $q$  is not reset, all traces in  $B_3$  satisfy  $q$ , so if  $B = B_3$  then  $\llbracket K_1 q \rrbracket_B = \mathcal{G}^+$ . Observe that there are traces that satisfy  $M_2 p$  under  $B_3$ , but that are not in  $B_\flat$ . However, if  $B = B_3$  then  $B_\flat \subseteq \llbracket M_2 p \rrbracket_B$ . Therefore, agent 2 can set  $q$  to zero under  $B_3$ , so  $B_4 = F(B_3) = B_\flat$ .

If  $B = B_4$  then  $\llbracket K_1 q \rrbracket_B = \emptyset$  and  $B_\flat \cap \llbracket M_2 p \rrbracket_B = B_\flat$ . Only agent 2 can act, therefore  $B_5 = F(B_4)$  is equal to  $B_1$ . At this point the iteration sequence repeats. The sequence is drawn in figure 4.

In this example the semantics as defined in (11) yields  $B_2$  as the semantics of the program, since  $B_1 \supseteq B_2 \not\supseteq B_3$ . Therefore, the semantics is that the system remains in its initial state.

#### 5.4. An infinite iteration sequence

In this example we show that the iteration sequence may become infinite. There are again two agents 1 and 2. Agent 1 has private integer variables  $p$  and  $m$ , and agent 2 has private integer variables  $q$  and  $n$ . Initially  $p = q = 0$  and  $m = n = 1$ . Agent 1 may always increment  $m$ . If agent 1 knows that  $q$  is



0, she can set  $p$  to 1. If agent 1 considers it possible that  $q$  is between 1 and  $m$ , she can advance  $p$  to  $m + 1$ . Agent 2 can do similar actions. If we allow equalities and inequalities to appear in the epistemic formulas, then the KBP is

$$\begin{aligned} & ( \quad \Lambda \\ & \cup \quad m := m + 1 \\ & \cup \quad n := n + 1 \\ & \cup \quad K_1(q = 0) ? ; p := 1 \\ & \cup \quad K_2(p = 0) ? ; q := 1 \\ & \cup \quad M_1(1 \leq q \leq m) ? ; p := m + 1 \\ & \cup \quad M_2(1 \leq p \leq n) ? ; q := n + 1 \\ & )^*. \end{aligned}$$

Again  $B_0 = \mathcal{G}^+$ . When  $\mathcal{G}^+$  is considered, agent 1 can set  $p := m + 1$ . Since agent 1 cannot inspect  $q$ , she can always consider a trace in  $\mathcal{G}^+$  possible such that  $1 \leq q \leq m$ . Agent 1 cannot set  $p$  to 1, because for  $B = \mathcal{G}^+$  we have  $\llbracket K_1(q = 0) \rrbracket_B = \emptyset$ . Agent 1 can always increment  $m$ . Similar arguments for agent 2 show that she cannot set  $q$  to 1. Therefore,  $B_1 = F(B_0)$  consists of the traces of program

$$( \Lambda \cup m := m + 1 \cup n := n + 1 \cup p := m + 1 \cup q := n + 1 )^*.$$

All traces in  $B_1$  satisfy  $p \neq 1 \neq q$ . So in  $B_1$  agent 1 only considers  $1 \leq q \leq m$  possible if  $m \geq 2$ . Again agent 1 never knows that  $q = 0$ . Similar arguments hold for agent 2. Therefore,  $B_2 = F(B_1)$  consists of the traces of program

$$( \Lambda \cup m := m + 1 \cup n := n + 1 \cup (m \geq 2) ? ; p := m + 1 \cup (n \geq 2) ? ; q := n + 1 )^*.$$

All traces in  $B_2$  satisfy  $p \notin \{1, 2\}$  and  $q \notin \{1, 2\}$ . In  $B_2$  agent 1 only considers  $1 \leq q \leq m$  possible if  $m \geq 3$ , but she never knows that  $q = 0$ . Continuing the same reasoning as above, we observe that for all integers  $k \geq 1$ , all traces in  $B_{k+1} = F(B_k)$  satisfy  $p \notin \{1, \dots, k\}$  and  $q \notin \{1, \dots, k\}$ . Furthermore, it is not hard to observe that  $B_{k+1} \subseteq B_k$ . Therefore, the transfinite iteration is  $B_\omega = \bigcap_{k < \omega} \bigcup_{k \leq l < \omega} B_l = \bigcap_{k < \omega} B_k$ . In  $B_\omega$  values of  $p$  and  $q$  are never increased, that is  $B_\omega$  consists of traces of program

$$( \Lambda \cup m := m + 1 \cup n := n + 1 )^*.$$

All traces in  $B_\omega$  satisfy  $p = 0 = q$ , so in  $B_\omega$  agent 1 knows that  $q = 0$  and can thus set  $p$  to 1. However,  $p$  cannot be incremented further as there are no traces in  $B_\omega$  for which  $1 \leq q \leq m$  holds for any  $m$ . Similarly,  $q$  can only be set to 1. So  $B_{\omega+1}$  consists of traces of program

$$( \Lambda \cup m := m + 1 \cup n := n + 1 \cup p := 1 \cup q := 1 )^*.$$

In  $B_{\omega+1}$ , agent 1 never knows that  $q = 0$ . However, she may consider it possible that agent 2 has incremented  $q$  such that  $1 \leq q \leq m$ , so she can set  $p$  to  $m + 1$ . Therefore,  $B_{\omega+2} = F(B_{\omega+1})$  is equal to  $B_1$ , so the iteration sequence repeats itself. The sequence is drawn in figure 5.

In this example, the semantics defined in (11) yields  $B_\omega$  as the semantics of the program, since the sequence  $B_1, B_2, \dots, B_\omega$  is the longest decreasing sequence starting with  $B_1 = B_{\omega+2}$ . In this program agents cannot make any assumption on the values of private variables of the other agent. The values of  $p$  and  $q$  should remain unchanged, while  $m$  and  $n$  may increase.

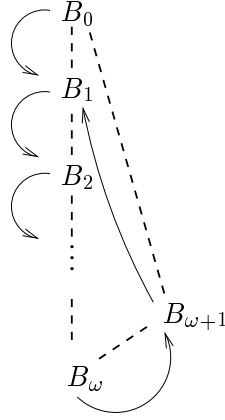


Figure 5. Example 5.4: an infinite sequence of iterations

### 5.5. Message passing

Message passing can be modelled by means of a variable that can be written by one agent and read by another. In this example, the agents 1 and 2 communicate via the integer variable  $q$ . Agent 1 has a private integer variable  $p$  and agent 2 has a private integer variable  $r$ .

All variables are initially 0. The state space consists of triples  $(p, q, r)$ , and we have

$$\begin{aligned} (p, q, r) \sim_1 (p', q', r') &\Leftrightarrow p = p' \wedge q = q', \\ (p, q, r) \sim_2 (p', q', r') &\Leftrightarrow q = q' \wedge r = r'. \end{aligned}$$

Agent 1 may increment  $p$  when she considers it possible that  $p \leq r$ . She may increment  $q$  when she knows that  $q < p$ . Agent 2 may increment  $r$  when she knows that  $r < p$ . The program is

$$\begin{aligned} Pg = & ( \Lambda \\ & \cup M_1 (p \leq r) ? ; p := p + 1 \\ & \cup (q < p) ? ; q := q + 1 \\ & \cup K_2 (r < p) ? ; r := r + 1 \\ & )^* . \end{aligned}$$

Because agent 1 can read both  $p$  and  $q$ , the test  $q < p$  is equivalent to  $K_1(q < p)$ .

Intuitively, one could expect the following behaviours from this program. Agent 1 may increment  $p$ , followed by  $q$ . Then agent 2 may increment  $r$ . Since agent 1 knows this, she may increment  $p$  after the incrementation of  $q$ ; she need not wait for the incrementation of  $r$ , which is hidden to her anyhow. In appendix A.1 the iteration sequence, sketched in figure 6, is worked out. As seen in figure 6, a fixpoint occurs. It turns out that this fixpoint corresponds with our intuition.

### 5.6. The Unexpected Hanging Paradox

The last example is inspired on a well-known paradox, first mentioned in [16] as the case of the “Class A blackout”. Presently, it is commonly known as “The Surprise Examination”, or “The Unexpected Hanging”.

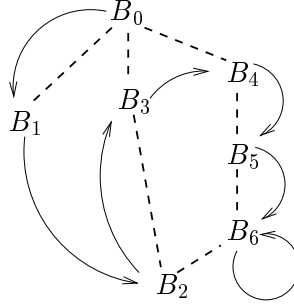


Figure 6. Example 5.5: nonmonotonic iteration that leads to a fixpoint.

In the unexpected hanging paradox, a convicted prisoner is to be executed at noon within seven days (numbered 0 to 6), but the judge tells him that he will not know the day of his execution, even on the beginning of that day itself. The prisoner might then reason that he cannot be executed on day 6, because when he would still be alive at the beginning of day 6, he would know that he would be executed that day. By backward induction he might reason that he cannot be executed without knowing that he will be executed. Yet, on day 2 the prisoner is surprised to meet the executioner.

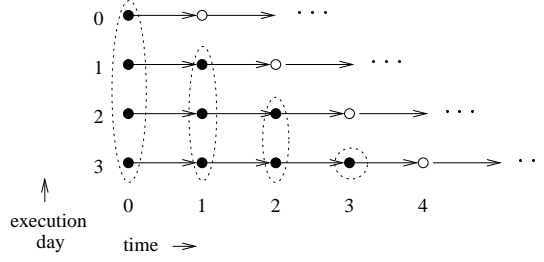
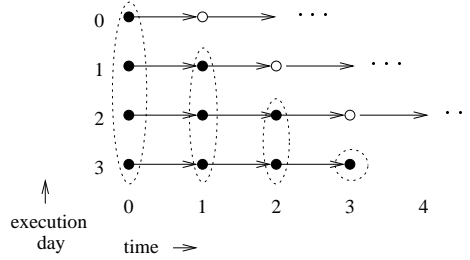
Let us formulate this situation as a KBP. The state space consists of three variables: an integer *day*, the day of execution *exec*  $\in \{0, \dots, 6\}$  and a boolean *dead*. Initially *day* = 0, *dead* = 0 and the precise day *exec* of execution is unknown to the agent. The agent only knows that *exec* lies in the range  $\{0, \dots, 6\}$ . The convicted agent 1 can observe *day* and *dead*, but not *exec*. The program is

$$\begin{aligned}
 & ( \quad \Lambda \\
 & \cup \quad (day = exec \wedge \neg K_1(day = exec) \wedge \neg dead) ? ; dead := 1 \\
 & \cup \quad (day \neq exec \vee dead) ? ; day := day + 1 \\
 & )^*
 \end{aligned}$$

In our model the program has an execution with *dead* being set when *exec* < 6. If initially *exec* = 6, the agent will know this at *day* = 6. So at that time, the program is stuck at day six. This seems to comply with intuition.

We illustrate the iterations  $B_1$  and  $B_2$  in a smaller version of this paradox, where the execution day lies in the range  $\{0, \dots, 3\}$ . The traces of the iterations  $B_1$  and  $B_2$  are given in figures 7 and 8. The arrows corresponding to  $\Lambda$  are not shown. The dashed ellipses give the uncertainty of the agent, and are induced by  $\sim_1$ . The filled bullets are the states where the agent is alive, and the open bullets are the states where the agent has been executed. In  $B_1 = F(B_0)$ , the agent has no knowledge on the execution day, so he can be executed on all days. In  $B_2 = F(B_1)$ , the agent cannot be executed on day 3, but he can be executed on the earlier days. The iteration  $B_2$  is a fixpoint of  $F$ , and is chosen as the semantics of the program.

Technically, our formalism models deadlock by forced stutterings, since the alternative  $\Lambda$  can always be executed. Moreover, no fairness assumptions are given, e.g. there is no guarantee that time progresses. Therefore, the agent cannot use backward induction to exclude execution sequences that lead to deadlock. This resolves the paradox. This is in agreement with the philosophical analysis based on dynamic

Figure 7. Traces of  $B_1 = F(B_0)$  for  $exec \in \{0, \dots, 3\}$ Figure 8. Traces of  $B_2 = F(B_1)$  for  $exec \in \{0, \dots, 3\}$ 

epistemic logic given in [10].

## 6. Conclusions and directions for future research

We have presented an asynchronous version of knowledge-based programming based on stutterings and interleaving semantics, that can be compared with the framework of systems and runs of Fagin et al. in [8]. The fixpoint equation that the semantics of asynchronous KBP should satisfy, does not always have a solution. Instead of restricting the semantics definition to the programs that lead to a fixpoint equation with a unique solution, we proposed a method to assign semantics to all KBPs in our framework, by calculating a sequence of iterations. If this sequence ends in a fixpoint, that fixpoint is taken to be the semantics of the KBP. If not, we choose a specific iteration as the semantics. To justify this choice, a number of examples have been worked out.

We have not found a suitable subclass of KBPs for which the automorphism  $F$  is monotonic. It would be useful to study what conditions, if any, would guarantee monotonicity of  $F$ .

The idea to approximate the semantics via sets  $B_\lambda$  of traces under consideration came up while trying to attach meanings to the example programs of Section 5. Our semantics definition is a proposal: one could advocate other choices within the same or similar sequences of iterations, such as the re-entry iteration (i.e. the first iteration  $B_\kappa$  with  $B_\kappa = B_{\kappa+\lambda}$  for some  $\lambda > 0$ ) or the first local minimum (i.e. the first iteration  $B_\kappa$  for which  $B_{\kappa+1} \not\subseteq B_\kappa$ ). The re-entry semantics gives a different meaning than our proposed semantics for the example of Section 5.4, and the semantics of the first local minimum differs from our proposed semantics in the example of Section 5.5. We find our semantics more intuitively

appealing than the re-entry or first local minimum semantics, but it is very well possible that our choice can be improved. This may especially be the case when attention is restricted to an interesting subclass of asynchronous KBPs.

We have not looked at simplifications of our semantics that would make model checking feasible. In general, we think that reasoning about a system during design (by using proof rules) is to be preferred above reasoning afterwards. As a consequence, our main concern is the development of useful proof rules. We must admit, however, to fear that the current semantics does not admit very useful proof rules, and it remains to be seen whether alternative and more useful semantics can be found. The relation with the predicate transformer approach of [20] could be explored in more detail.

In our current framework we do not consider temporal formulas. No fairness assumptions or progress properties can be expressed. It would be interesting to extend the framework to infinite traces (runs) and progress properties.

Another direction of research would be to study refinement relations between KBPs in our framework. A KBP can then gradually be refined to a knowledge-free program. This would eliminate the need to explicitly calculate an iteration sequence.

It may be useful to express the automorphism  $F$  defined in (10) in terms of automata. The set of traces under consideration can be defined as finite state automaton that recognizes traces from that set. An automaton transformer is then associated with our function  $F$ . This automaton transformer might be studied in a different setting.

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## A. Appendix

### A.1. Message Passing

In this section we work out the example from Section 5.5. Recall that there are two agents 1 and 2, and three integer variable  $p$ ,  $q$  and  $r$  which are all initially 0. Variable  $q$  is shared,  $p$  is private to agent 1 and  $r$  is private to agent 2. So the state space consists of triples  $(p, q, r)$ , and we have

$$\begin{aligned} (p, q, r) \sim_1 (p', q', r') &\Leftrightarrow p = p' \wedge q = q', \\ (p, q, r) \sim_2 (p', q', r') &\Leftrightarrow q = q' \wedge r = r'. \end{aligned}$$

The program is

$$\begin{aligned} Pg = & \left( \begin{aligned} &\Lambda \\ &\cup M_1 (p \leq r) ? ; p := p + 1 \\ &\cup \quad (q < p) ? ; q := q + 1 \\ &\cup K_2 (r < p) ? ; r := r + 1 \end{aligned} \right)^* . \end{aligned}$$

Agent 2 can infer  $r < p$  from  $r < q$ , since she can read  $q$  and agent 1 preserves the invariant  $q \leq p$ . Operationally speaking, agent 1 uses variable  $q$  as a message to agent 2 that  $p$  has been incremented.

In  $B_0 = \mathcal{G}^+$ , agent 1 considers it possible that  $p \leq r$ , but agent 2 has no knowledge whether  $r < p$ . Therefore the traces in  $B_1 = F(B_0)$  are the traces generated by the knowledge-free program  $P1$

$$P1 = ( \Lambda \\ \cup (p \leq r) ? ; p := p + 1 \\ \cup (q < p) ? ; q := q + 1 \\ )^* .$$

For the next iterations it is useful to only consider traces that can possibly be generated by program  $Pg$ . Therefore, while determining the set of traces that satisfy  $K_2(r < p)$  given a set  $B$ , attention can be restricted to traces in the set  $B_b$ , which consist of traces generated by the flat, guardless program

$$( \Lambda \cup p := p + 1 \cup q := q + 1 \cup r := r + 1 )^* .$$

For  $B = B_1$ , we have that  $\llbracket M_1(p \leq r) \rrbracket_B$  consists of traces  $p \leq 0$ , since  $r$  remains 0 in  $B_1$ . On the other hand,  $B_b \cap \llbracket K_2(r < p) \rrbracket_B$  consists of traces of  $B_b$  that end with  $r < q$ , since  $q \leq p$  is invariant in  $B_1$ . Therefore,  $B_2 = F(B_1)$  consists of traces generated by

$$P2 = ( \Lambda \\ \cup (p \leq 0) ? ; p := p + 1 \\ \cup (q < p) ? ; q := q + 1 \\ \cup (r < q) ? ; r := r + 1 \\ )^* .$$

Note that the traces in  $B_2$  satisfy the invariant  $0 \leq r \leq q \leq p \leq 1$ .

For  $B = B_2$ , we have that  $\llbracket M_1(p \leq r) \rrbracket_B$  consists of traces with  $p \leq q \wedge p \leq 1$ , since  $r \leq 1$  holds in  $B_1$ . On the other hand,  $B_b \cap \llbracket K_2(r < p) \rrbracket_B$  consists of the traces of  $B_b$  that end with  $r < q \vee 2 \leq q$ . In fact, no trace  $xs$  with  $2 \leq q$  is indistinguishable for 2 from any trace in  $B_2$ , since all traces in  $B_2$  satisfy  $q \leq 1$ . It follows that  $B_3 = F(B_2)$  consists of traces generated by

$$P3 = ( \Lambda \\ \cup (p \leq q \wedge p \leq 1) ? ; p := p + 1 \\ \cup (q < p) ? ; q := q + 1 \\ \cup (r < q \vee 2 \leq q) ? ; r := r + 1 \\ )^* .$$

Note that traces in  $B_3$  satisfy the invariants  $0 \leq r$  and  $0 \leq q \leq p \leq 2$  and  $q < 2 \Rightarrow r \leq q$ .

For  $B = B_3$ , we have that  $\llbracket M_1(p \leq r) \rrbracket_B$  consists of traces with  $p \leq q \vee 2 \leq q$ , since in  $B_3$ ,  $r$  is only bounded by  $q$  while  $q < 2$ . On the other hand,  $B_b \cap \llbracket K_2(r < p) \rrbracket_B$  consists of traces of  $B_b$  that end with  $r < q \vee 3 \leq q$ . No trace satisfying  $3 \leq q$  is indistinguishable for 2 from any trace in  $B_3$ , since traces in  $B_3$  satisfy  $q \leq 2$ . It follows that  $B_4 = F(B_3)$  consists of the traces generated by

$$P4 = ( \Lambda \\ \cup (p \leq q \vee 2 \leq q) ? ; p := p + 1 \\ \cup (q < p) ? ; q := q + 1 \\ \cup (r < q \vee 3 \leq q) ? ; r := r + 1 \\ )^* .$$

Observe that traces in  $B_4$  satisfy the invariants  $0 \leq r$  and  $0 \leq q \leq p$  and  $q < 3 \Rightarrow r \leq q$ .

For  $B = B_4$ , we have that  $\llbracket M_1(p \leq r) \rrbracket_B$  consists of traces with  $p \leq q \vee 3 \leq q$ , since in  $B_4$ ,  $r$  is only bounded by  $q$  while  $q < 3$ . On the other hand,  $B_b \cap \llbracket K_2(r < p) \rrbracket_B$  consists of those traces of  $B_b$  that end with  $r < q$  because of the invariant  $q \leq p$  in  $B_4$ . It follows that  $B_5 = F(B_4)$  consists of the traces generated by

$$\begin{aligned} \text{P5} = & ( \ \Lambda \\ & \cup (p \leq q \vee 3 \leq q) ? ; p := p + 1 \\ & \cup (q < p) ? ; q := q + 1 \\ & \cup (r < q) ? ; r := r + 1 \\ & )^* . \end{aligned}$$

Traces in  $B_5$  satisfy the invariant  $0 \leq r \leq q \leq p$ .

For  $B = B_5$ , we have that  $\llbracket M_1(p \leq r) \rrbracket_B$  consists of the traces with  $p \leq q$  since  $r$  is only bounded by  $q$  in  $B$ . On the other hand,  $B_b \cap \llbracket K_2(r < p) \rrbracket_B$  consists of those traces of  $B_b$  that end with  $r < q$  because all traces in  $B_5$  satisfy the invariant  $q \leq p$ . It follows that  $B_6 = F(B_5)$  consists of traces generated by

$$\begin{aligned} \text{P6} = & ( \ \Lambda \\ & \cup (p \leq q) ? ; p := p + 1 \\ & \cup (q < p) ? ; q := q + 1 \\ & \cup (r < q) ? ; r := r + 1 \\ & )^* . \end{aligned}$$

Note that traces in  $B_6$  satisfy the invariant  $0 \leq r \leq q \leq p \leq q + 1$ .

Similarly as before, for  $B = B_6$ , we have that  $\llbracket M_1(p \leq r) \rrbracket_B$  consists of traces with  $p \leq q$  since  $r$  is only bounded by  $q$  in  $B$ . On the other hand,  $B_b \cap \llbracket K_2(r < p) \rrbracket_B$  consists of those traces of  $B_b$  that end with  $r < q$  because of the invariant  $q \leq p$  in  $B_6$ . It follows, that  $F(B_6) = B_6$ , a fixpoint. Moreover, this fixpoint corresponds with the intuition sketched in Section 5.5 and is chosen as the semantics defined in (11). The iteration sequence is shown in figure 6.